# AI, Data Science & Health

Elaine Nsoesie, PhD Assistant Professor of Global Health Email: onelaine@bu.edu / Twitter: @ensoesie



How will you use Al and data science to improve health in Africa?

## Health in Sub-Saharan Africa

- Life expectancy has increased from 53 years in 2000 to 64 years in 2017
  - Men is 62, Women is 66
- The number of children dying before the age five has decreased from 45% of all deaths in 1950 to 10% in 2017
- Overall decrease in mortality



## Health in Sub-Saharan Africa

Leading causes of death for adults (15-49 years)

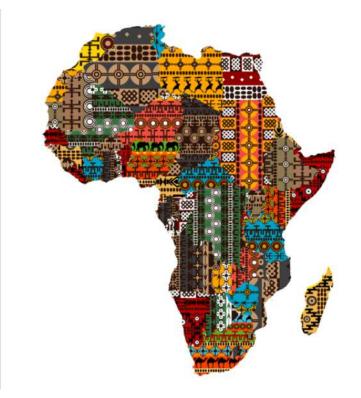
- AIDS
- Tuberculosis
- Malaria
- Maternal disorders
- Road injuries



## Health in Sub-Saharan Africa

Impending epidemics of noncommunicable diseases

- High blood pressure
- High blood sugar
- Obesity
- Alcohol use
- Diabetes
- Heart attacks
- Stroke



## Exercise

Write down one health problem that you can solve with AI or data science. (10 mins)

## For those of you interested, the goal is to

- Come up with a simple/complicated problem that we can solve with Al
- 2. Think about what data we would need to solve this problem
- 3. Think about what methods we can use to solve this problem and why these methods are appropriate
- 4. Come up with a research plan
- 5. If you want to pursue that research plan beyond this course, I would be happy to advise you

## Outline for the remainder of this session



## Data

## Types of digital data & uses

Mobile Apps & Crowdsourcing

Search

**Social** media

Consumer reviews

Remote sensing/place

**News** 













**DigitalGlobe** 



HealthMap













## Some links to digital data sources

- trends.google.com
- https://dataforgood.fb.com/
- developer.twitter.com
- https://developers.google.com/maps

## Surveys

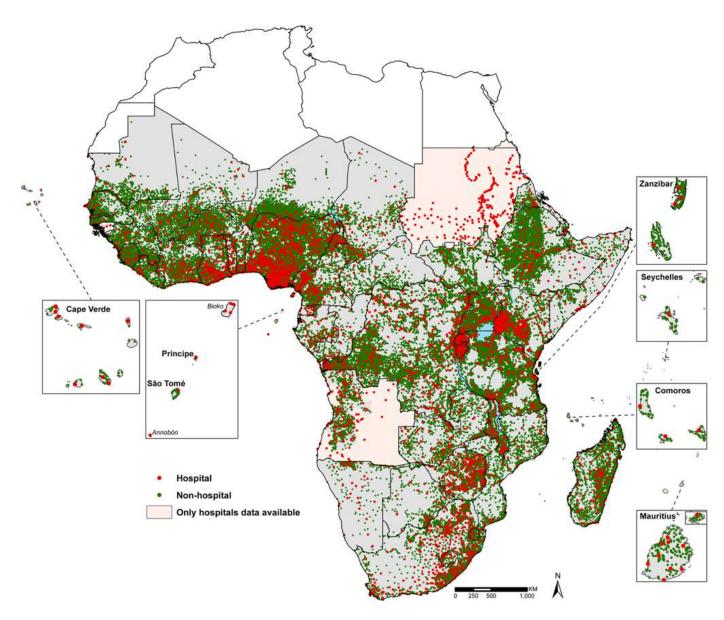
- Country health surveys (e.g., family health survey)
- Census
- UNICEF
- DHS
- World Bank
- NGOs
- Published research studies

# A spatial database of health facilities managed by the public health sector in sub Saharan Africa

Maina et al. (2019)

The distribution of 96,395 geocoded public health facilities in sub Saharan Africa. Red dots represent hospitals (n = 4,930) and green dots represent non-hospitals (n = 91,465). Facilities have been mapped on Global Administrative Unit Layers (GAUL) 2008 admin0 boundaries

(<a href="http://www.fao.org/geonetwork/srv/en/main.home">http://www.fao.org/geonetwork/srv/en/main.home</a>).









WHO WE ARE

WHAT WE DO

WHERE WE WORK

DATA

**PUBLICATIONS** 

**TOPICS** 

The DHS Program > What-We-Do > Survey Types > Demographic and Health Survey (DHS)

#### What We Do

SURVEY TYPES

Demographic & Health Survey (DHS)

AIDS Indicator Survey (AIS)

Service Provision Assessment (SPA)

Malaria Indicator Survey (MIS)

Key Indicators Survey (KIS)

Other Quantitative Surveys

Qualitative Research

#### **DHS** Overview

Demographic and Health Surveys (DHS) are nationally-representative household surveys that provide data for a wide range of monitoring and impact evaluation indicators in the areas of population, health, and nutrition.

#### **DHS Survey Types**

There are two main types of DHS Surveys:

- Standard DHS Surveys have large sample sizes (usually between 5,000 and 30,000 households) and typically are conducted about every 5 years, to allow comparisons over time.
- Interim DHS Surveys focus on the collection of information on key performance monitoring indicators but may not include data for all impact evaluation measures (such as mortality rates). These surveys are conducted between rounds of DHS surveys and have shorter questionnaires than DHS surveys. Although nationally representative, these surveys generally have smaller samples than DHS surveys.

#### **DHS Survey Topics**

#### **DHS** Resources

**OVERVIEW** 

METHODOLOGY

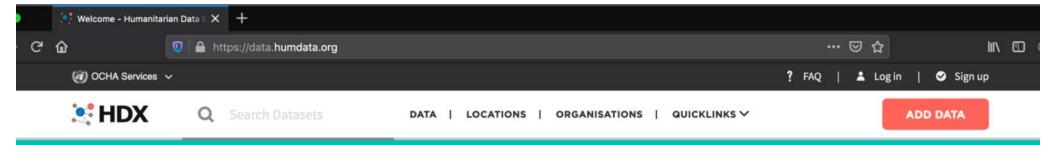
QUESTIONNAIRES

MANUALS

#### **DHS Survey Topics**

Information is available for the following topics, among others:

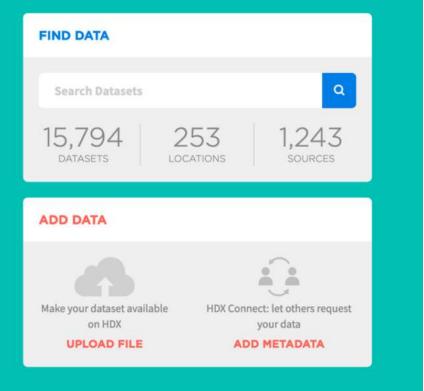
- Anemia prevalence of anemia, iron supplementation
- Child Health vaccinations, childhood illness, newborn care
- <u>Domestic Violence</u> (module) prevalence of domestic violence and consequences of violence
- Education literacy, attendance, highest level achieved
- · Environmental Health water, sanitation, cooking fuel
- Family Planning knowledge and use of contraceptives
- Female Genital Cutting (module) prevalence of and attitudes about female genital cutting
- Fertility and Fertility Preferences total fertility rate, desired family size, marriage and sexual activity
- · Gender/Domestic Violence history of domestic violence, frequency and consequences of violence
- <u>HIV/AIDS Knowledge, Attitudes, and Behavior</u> knowledge of HIV prevention, misconceptions, stigma, higher-risk sexual behavior, previous HIV testing
- HIV Prevalence Prevalence of HIV by demographic and behavioral characteristics
- Household and Respondent Characteristics electricity, housing quality, possessions, education and school attendance, age, sex, employment
- Infant and Child Mortality infant and child mortality rates
- Malaria ownership and use of mosquito nets, prevalence and treatment of fever, indoor residual spraying for mosquitoes
- Maternal Health antenatal, delivery and postnatal care
- Maternal Mortality (module)- maternal mortality ratio
- · Nutrition child feeding practices, vitamin supplementation, anthropometry, anemia, salt iodization
- Tobacco Use tobacco use, exposure to second-hand smoke
- Unmet Need for family planning
- Wealth division of households into 5 wealth quintiles to show relationship between wealth, population and health indicators
- Women's Empowerment gender attitudes, women's decision making power, education and employment of men vs. women

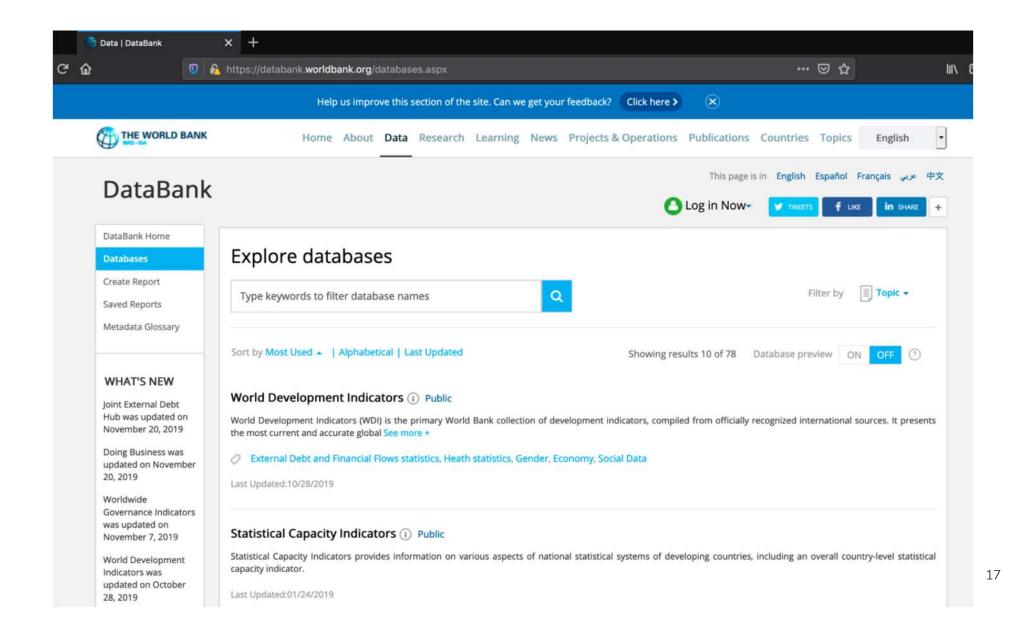


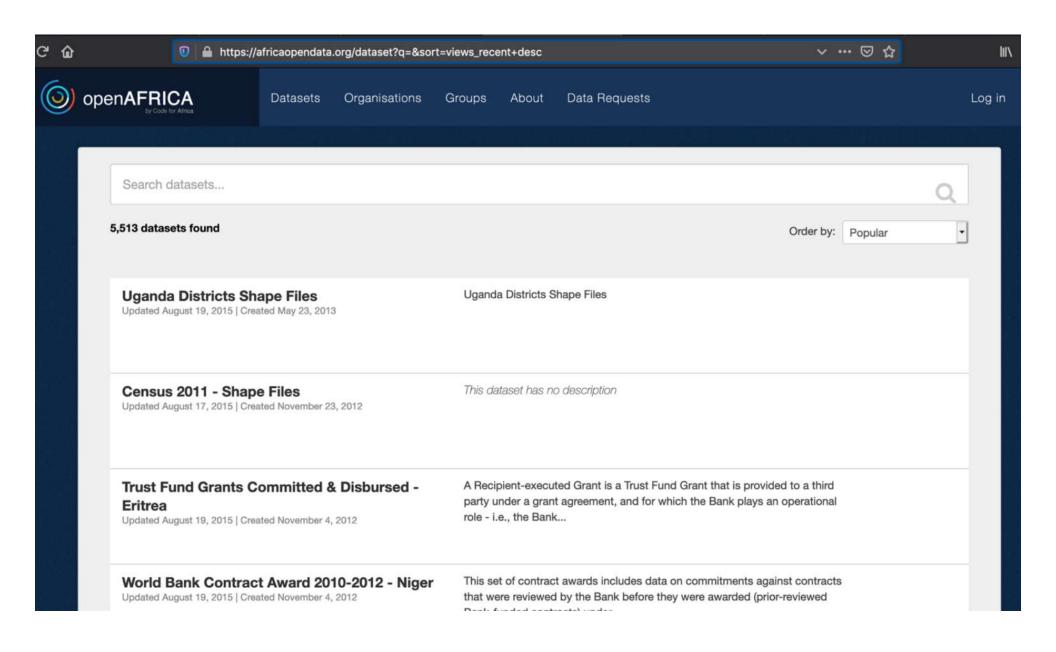
## The Humanitarian Data Exchange

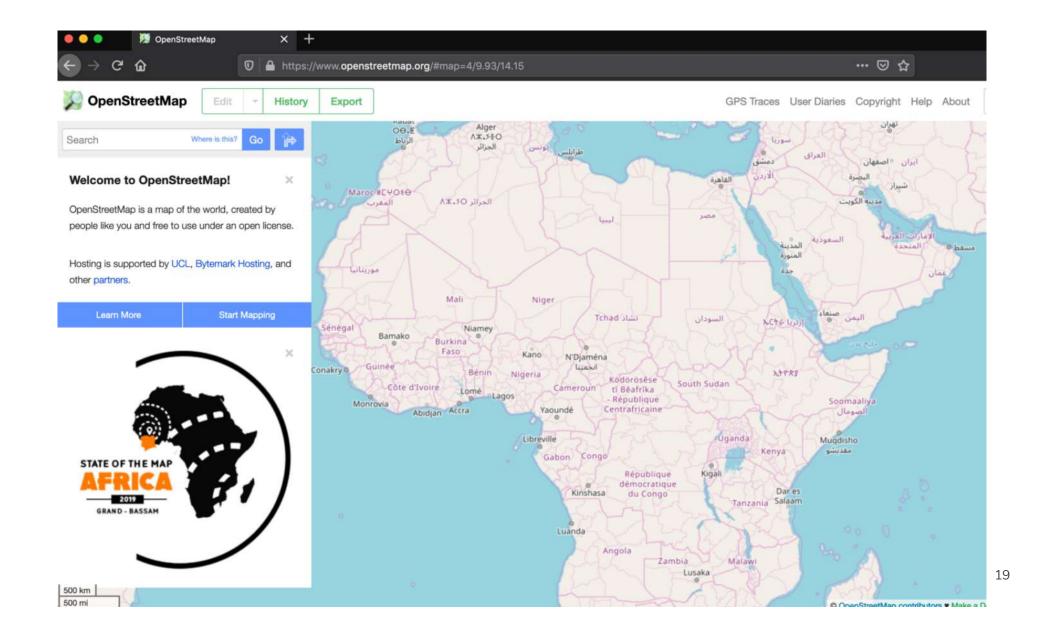
Find, share and use humanitarian data all in one place

**LEARN MORE** 













# The 2018 Nigeria HIV/AIDS Indicator and Impact Survey (NAIIS)

is a cross-sectional survey that assessed the prevalence of key human immunodeficiency virus (HIV)-related health indicators.











## **Nigeria Centre for Disease Control**

Protecting the health of Nigerians



About -

Publications -

Diseases -

News/Media

Training/Events -

**Projects** 

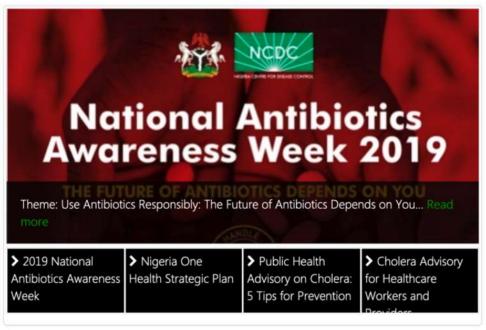
Jobs

Preparedness

Dashboard

Contact





## **Exercise**

What type(s) of data and what sources would you use to solve the health problem you previously identified? (10 mins)

## Methods

## Satellite Images and deep learning

About 55% of the world's population lives in urban areas. This number is expected to increase to 68% by 2050.



The field of urban health which started around the turn of the 21<sup>st</sup> century is focused on the study of how and why cities influence health.

### The foods we have access to







The foods we eat

## All neighborhoods are not created equal

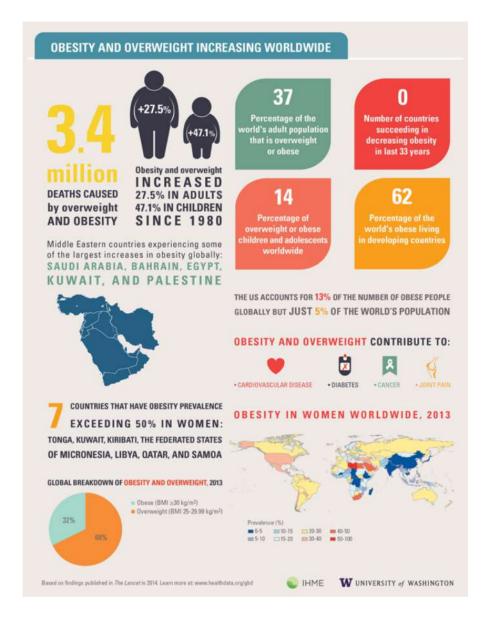




### Drivers of health in cities

The physical environment
The social environment
Access to health and social services

- Obesity is a complex health issue
- Multiple factors have been linked to obesity
- In the US, obesity affects about one-third of the adult population





# Many environmental factors have been linked to obesity

But there are <u>differences in measures and</u> <u>measurements</u> making it difficult to compare findings across cities

## Obesity prevalence across neighborhoods in six cities

- Memphis, Tennessee (6<sup>th</sup>)
- Estimates of obesity prevalence from CDC
- Seattle-Bellevue-Tacoma,
   Washington (32<sup>nd</sup>)
- Points of interest

San Antonio, Texas (8<sup>th</sup>)

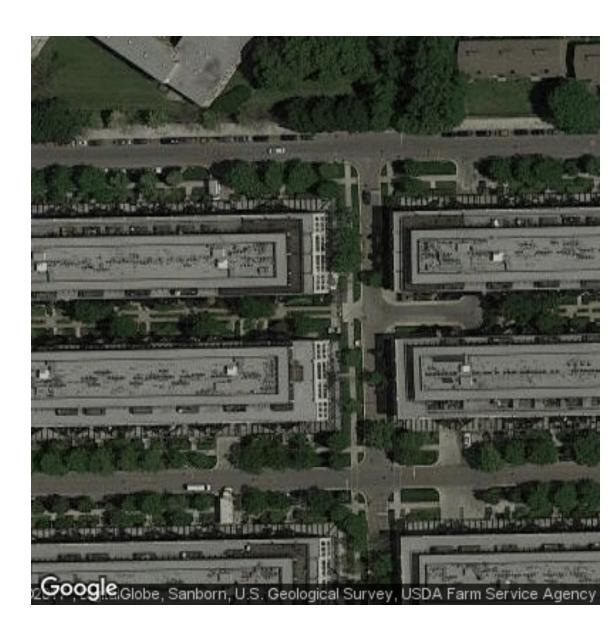
Per capita income

Los Angeles, CA (47<sup>th</sup>)

Satellite images from Google

~150,000 recent satellite images were downloaded from Google Static Maps API

Zoom level 8

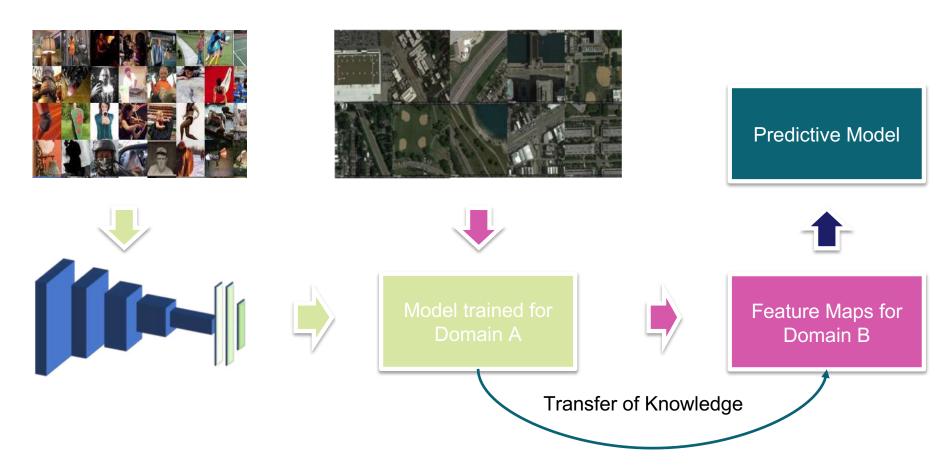


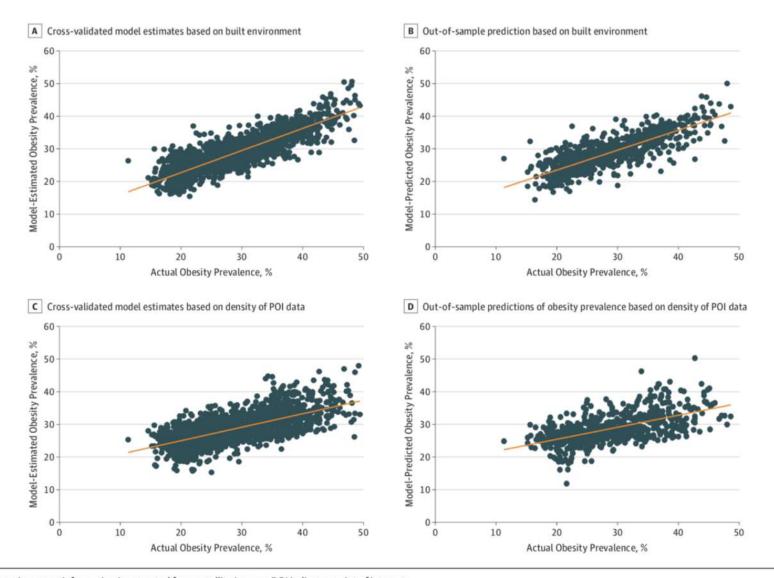
## Data - Satellite Images

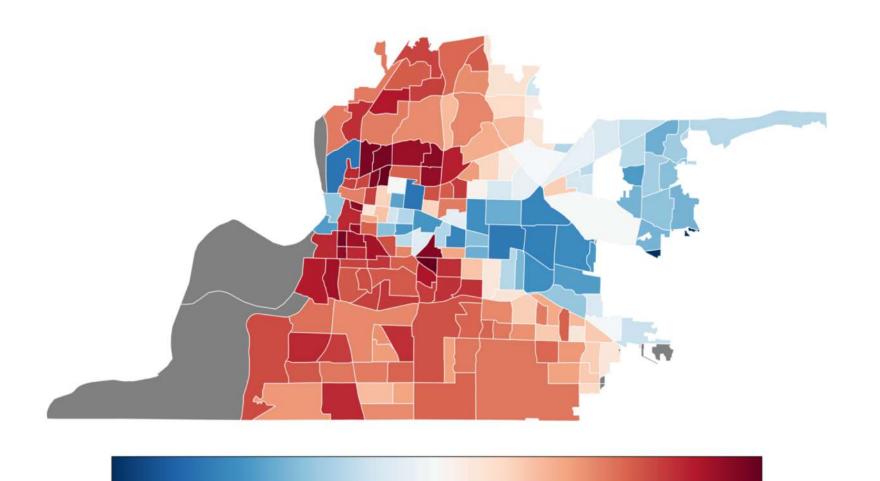




## Transfer Learning with VGG-CNN-F







30.6

34.4

38.1

41.9

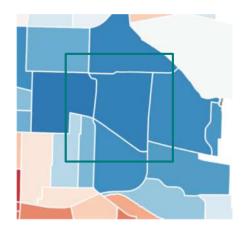
45.6

15.6

19.4

23.1

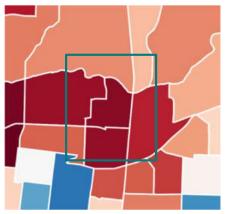
26.9







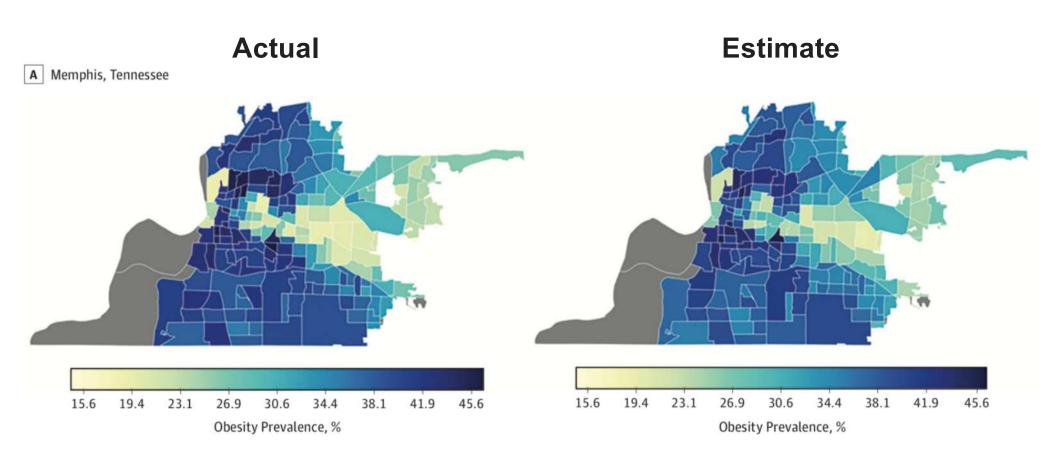












### Mapping

### Chikungunya Disease Transmission and Implications for Surveillance

## Chikungunya Disease Transmission and Implications for Surveillance

- Chikungunya virus was first identified in 1953 in Tanganyika (Tanzania)
- Chikungunya is an acute febrile illness
- Symptoms: incapacitating joint pain, high fever and skin rash
- Endemic in several Asian and African countries

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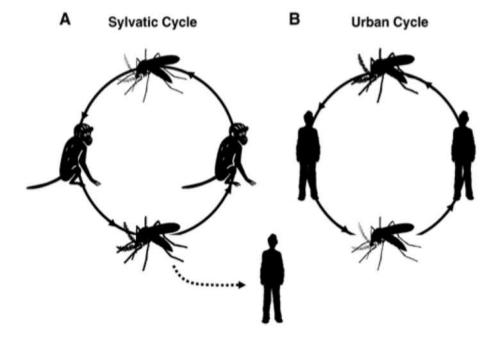
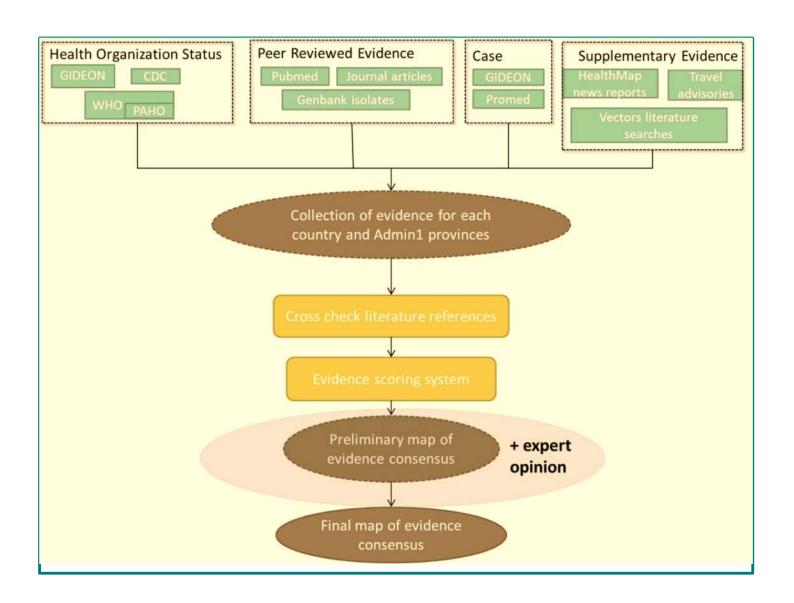


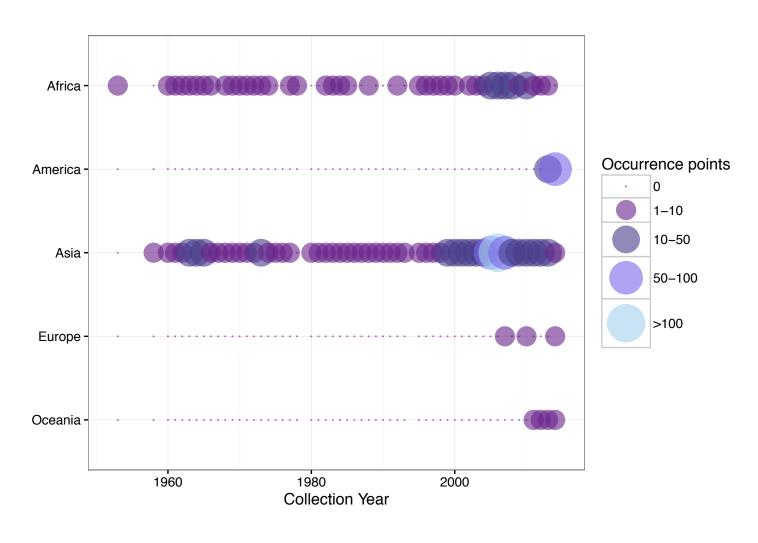
Fig 1. (A) CHIKV in Africa is maintained in a sylvatic cycle involving forest dwelling Aedes spp. mosquitoes and nonhuman primates. When sylvatic mosquito densities increase, often during periods of heavy rainfall, small human epidemics or sporadic human cases may occur. (B) In urban settings, CHIKV circulates in a man-mosquito-man cycle vectored principally by the anthropophilic A aegypti mosquito. Although A albopictus has been considered an accessory vector, some recent urban outbreaks have been vectored primarily by this mosquito.



### **Evidence Consensus**

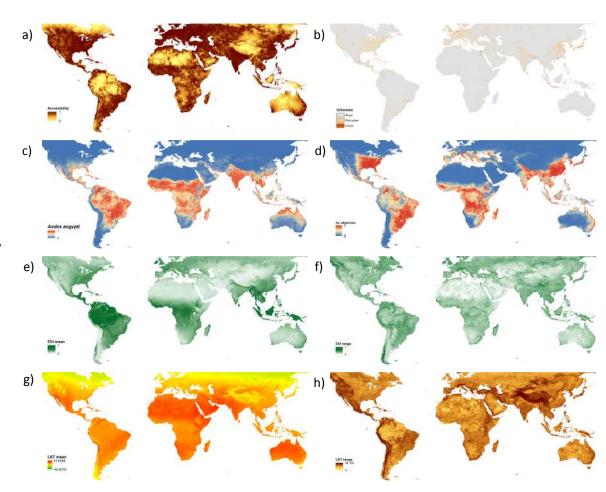
- Summary statistic representing confidence in presence or absence of Chikungunya for a given political region
- Evidence consensus score estimated at subnational level for France, Italy,
   India, Brazil, Mexico, Argentina, United States of America, and China
- Evidence consensus was mapped onto seven equidistant categories: complete presence/absence, good presence/absence, moderate presence/absence, and intermediate

### **Extraction of occurrence points**



## Extraction of covariates

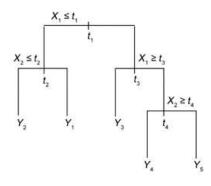
- a) Urban accessibility
- b) Urban, peri-urban and rural areas
- c) Aedes aegypti suitability
- d) Ae. albopictus suitability
- e) EVI mean values
- f) EVI range
- g) LST mean values
- h) LST range

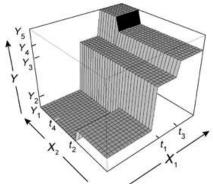


### **Model Specifications**

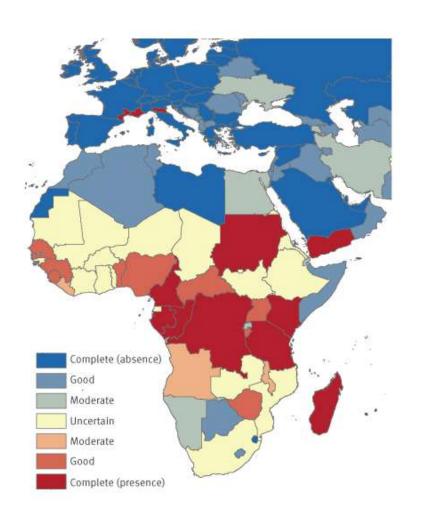
Boosted regression tree (BRT)

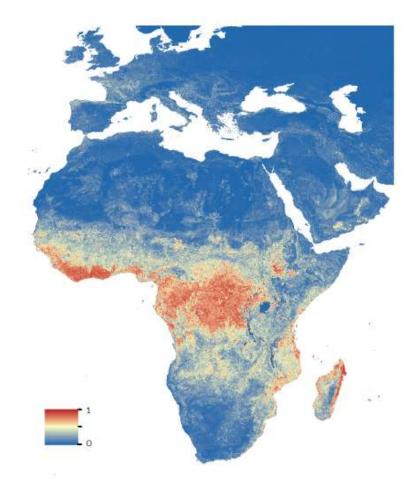
- Combining both, machine learning approaches and regression techniques
- Predictor variables can be of any type (numeric, binary, categorical, etc.)
- Boosting: a forward, stagewise procedure
- Outperforms MaxEnt, GARP, BIOCLIM

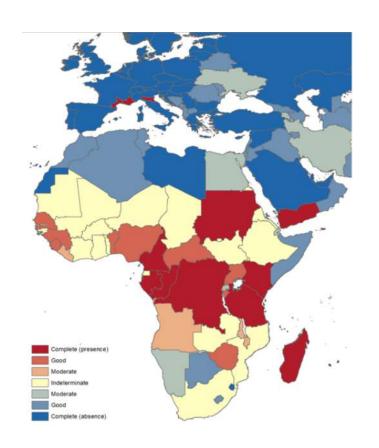












### Africa & Europe

54% of African countries had good or complete evidence for CHIKV transmission

Insufficient data to determine presence or absence for 30% of countries in Africa

Local transmission of CHIKV in Europe has been reported in Ravenna, northern Italy in 2007; the southeastern French city of Fréjus in 2010 and Montpellier, southern France in 2014

### Natural Language Processing

Open access Original article

### BMJ Open Sport & Exercise Medicine

### Social media captures demographic and regional physical activity

Nina Cesare, Quynh C Nguyen, Christan Grant, Elaine O Nsoesie 1,2

**To cite:** Cesare N, Nguyen QC, Grant C, et al. Social media captures demographic and regional physical activity. BMJ Open Sport & Exercise Medicine 2019;**5**:e000567. doi:10.1136/ bmjsem-2019-000567

► Additional material is published online only. To view, please visit the journal online (http://dx.doi.org/10. 1136bmjsem-2019-000567).

Accepted 2 July 2019

#### ABSTRACT

**Objectives** We examined the use of data from social media for surveillance of physical activity prevalence in the USA.

Methods We obtained data from the social media site Twitter from April 2015 to March 2016. The data consisted of 1 382 284 geotagged physical activity tweets from 481 146 users (55.7% men and 44.3% women) in more than 2900 counties. We applied machine learning and statistical modelling to demonstrate sex and regional variations in preferred exercises, and assessed the association between reports of physical activity on Twitter and population-level inactivity prevalence from the US Centers for Disease Control and Prevention.

**Results** The association between physical inactivity tweet patterns and physical activity prevalence varied by sex and region. Walking was the most popular physical activity for both men and women across all regions

#### What are the findings?

- Men mentioned engaging in higher intensity physical activities than women, which agrees with previous studies suggesting that women are less likely to meet recommendations for aerobic physical activity.
- There were differences in the types of physical activities reported across the four US regions.

#### How might it impact clinical practice?

Differences in the types of physical activities reported across sex and regions in the US can encourage discussions between clinicians and patients regarding exercise choices for weight loss and cardiovascular health.

### **Demographics**

208.9 Million

Social media users in the US (2017)

91% Of 18-29 year olds use Youtube 78%
Of 30-49 year olds use Facebook

20%
Of users
making <\$30k
use Twitter

78%
Of Hispanics use Youtube

Social media provides a window into the built environment, interactions, access, likes/dislikes, lifestyle, behaviors, choices

Social media can provide timely and low-cost information on community health



#jogging
3,125,055 posts















### Social media can provide data on

1. Activities

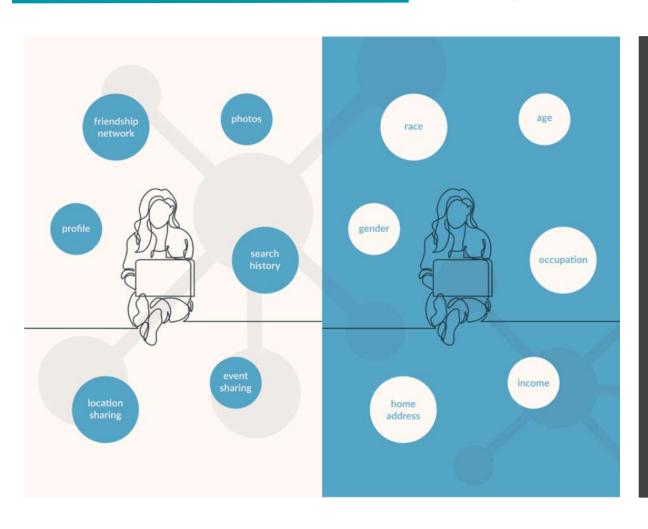
2. Sentiment

3. Scale

4. Changes

Does it matter that we do not have demographics from some social media platforms?

### **Understanding Demographic Disparities**



- Important gender differences in risk factors
- Important age differences in how people frame and discuss risk factors
- Spatial representation higher population, higher
   income counties typically
   have more data available

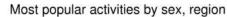
1,382,284

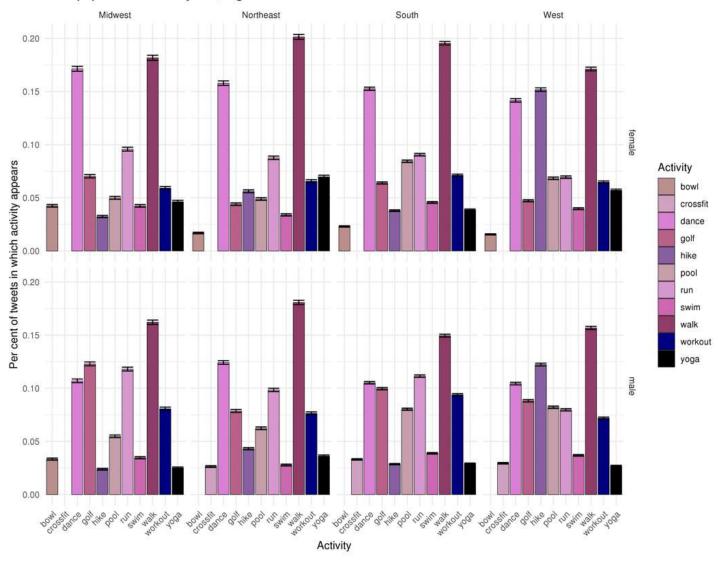
geotagged physical activity tweets

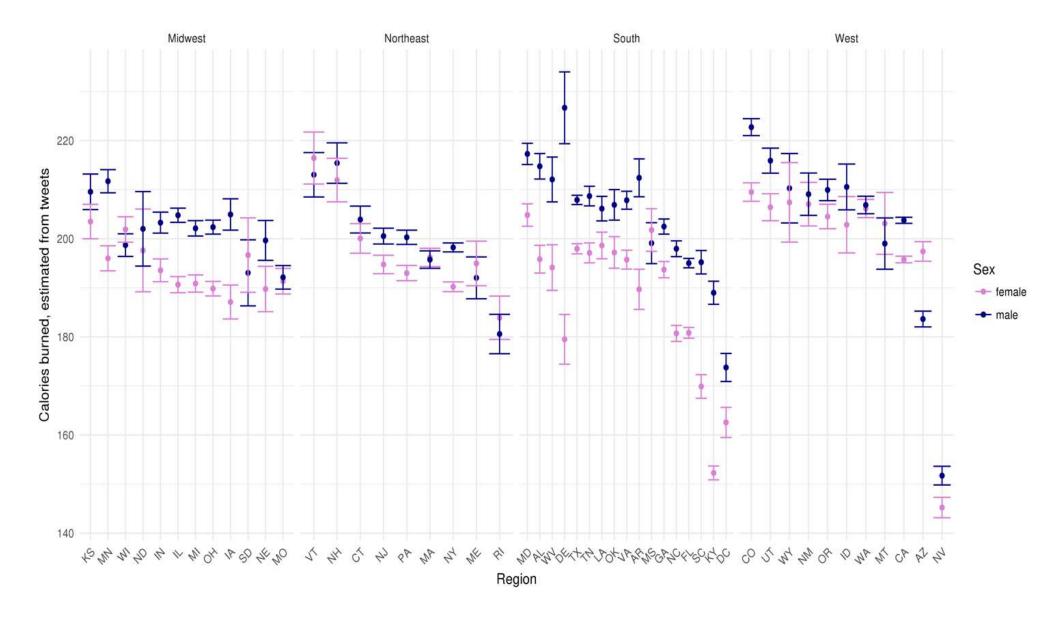
481,146 users

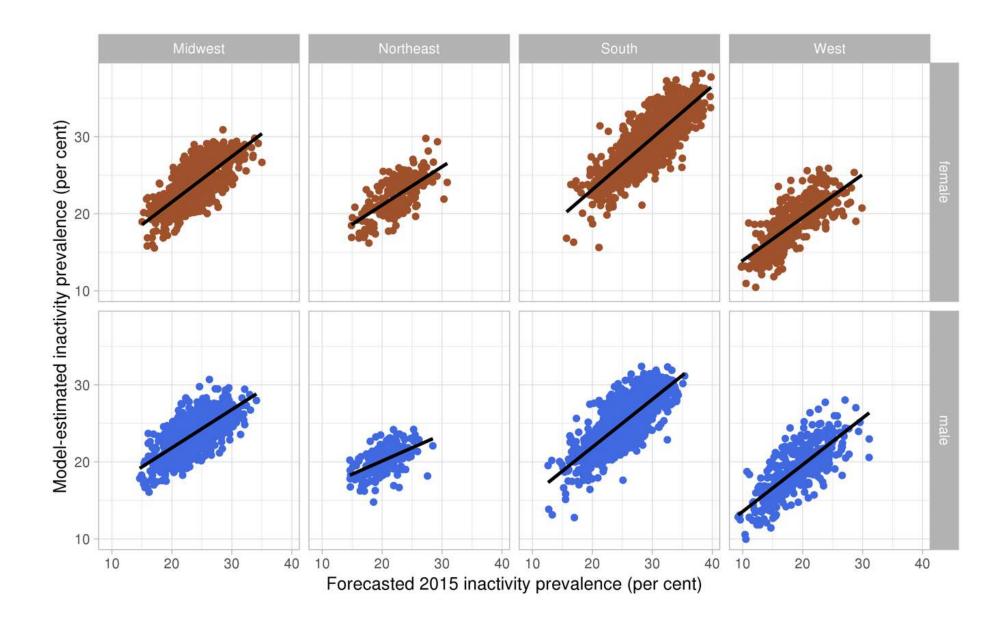
55.7% men and 44.3% women

> 2,900 counties









### Women's health

### Miscarriage

the spontaneous abortion of a viable foetus during the first 20 weeks of pregnancy



# Miscarriage facts and misinformation

Although an estimated 15 to 20% of known pregnancies end in a miscarriage, data suggest miscarriages are largely misunderstood and those affected can feel isolated.

# Miscarriage facts and misinformation

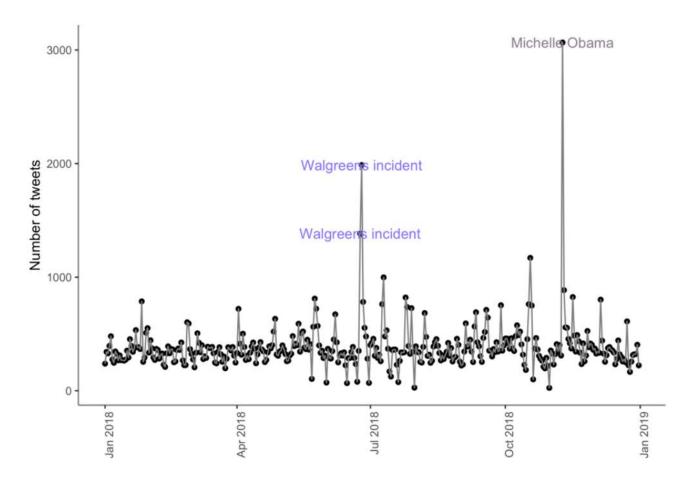
Most miscarriages are due to genetic problems that have nothing to do with witchcraft, environment or behavioral choices.

### **Data & Methods**

- Data from Twitter (291,443 postings)
- Applied Latent Dirichlet Allocation (LDA) to identify major topics of discussion
- Used Locally Weighted Least Squares Regression (LOESS) to identify trends and drivers of discussion
- Manually classified sentiment

### Topics of discussion

- Michelle Obama (8.4% of tweets)
- Celebrity (23.0%)
- Preterm birth (10.9%)
- Politics (17.6%)
- Loss and anxiety (10.1%)
- Ectopic pregnancy (7.50%)
- Healthcare (10.7%)
- Influenza vaccine (11.7%)



Discussions of personal experiences of miscarriage by celebrities such as, Michelle Obama can drive conversation on this topic

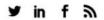
### References

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- Cesare N, Grant C, Hawkins JB, Brownstein JS, Nsoesie EO. Demographics in Social Media Data for Public Health Research: Does it matter? Bloomberg Data for Good Exchange. 2017 https://arxiv.org/abs/1710.11048
- Nsoesie EO\*\*, Kraemer MUG\*\* 1, Golding N, Pigott DM, Brady OJ, Moyes CL, Johansson MA, Gething PW, Velayudhan R, Khan K, Hay SI, Brownstein JS. Global Distribution and Environmental Suitability for Chikungunya Virus, 1952 to 2015. Eurosurveillance. 2016; 21(20):pii=30234. doi http://dx.doi.org/10.2807/1560- 7917.ES.2016.21.20.30234.

What methods would you use to solve the previously identified health problem? (10 mins)

### **Ethics**







### Can Social Media Predict When You'll Die?



FREEEDA/SHUTTERSTOCK.COM

By BOSTON UNIVERSITY // Futurity // OCTOBER 1, 2019

Could social media data predict your death?

SOCIAL MEDIA







Social media platforms like Twitter, Instagram, and

# Should your social media data be used to predict when you'll die?

Just because you can explore a research topic/question, doesn't mean you should.

Have your study participants consented to how you collect and use their data?

### Who is your data?

# Who benefits from your study?

10.28.19

## Technology biased against black patients runs rampant in hospitals

A new study shows that a widely used algorithm for predicting which patients get additional care is disproportionately counting out black patients—and could have left tens of thousands without adequate medical care.



# Is your data biased towards one group?

## Representation does not equal equity.

### Exercise

## What are the next steps for your project?

### Contributors Too many ©

### Thank you!

Contact

Email: onelaine@bu.edu

Twitter: @ensoesie